

published in final form at <https://doi.org/10.1111/aje.12553>. This article may be used for non-commercial purposes in accordance with Wiley Terms and Conditions for self-archiving.

## TITLE

Modelling large herbivore movement decisions: beyond food availability as a predictor of ranging patterns

## RUNNING TITLE

Modelling large herbivore movement decisions

## AUTHORS AND AFFILIATIONS

Victoria L. Boulton (*corresponding author*)<sup>a</sup>

- <sup>a</sup> University of Reading, School of Biological Sciences
- Victoria.Boulton@pgr.reading.ac.uk
- 105 Harborne Building, University of Reading, Whiteknights, Reading, Berkshire, RG6 6AS, United Kingdom
- Tel: +44(0)7772 654 616

Richard M. Sibly<sup>a</sup>

- <sup>a</sup> University of Reading, School of Biological Sciences
- r.m.sibly@reading.ac.uk

Tristan Quaife<sup>b</sup>

- <sup>b</sup> University of Reading, Department of Meteorology
- t.l.quaife@reading.ac.uk

Vicki Fishlock<sup>c, d</sup>

- <sup>c</sup> Amboseli Trust for Elephants & <sup>d</sup> Psychology, Faculty of Natural Sciences, University of Stirling
- vfishlock@elephanttrust.org

Cynthia Moss<sup>c</sup>

- <sup>c</sup> Amboseli Trust for Elephants
- cmoss@elephanttrust.org

Phyllis C. Lee<sup>d, c</sup>

- <sup>d</sup> Psychology, Faculty of Natural Sciences, University of Stirling & <sup>c</sup> Amboseli Trust for Elephants
- phyllis.lee@stir.ac.uk

## ABSTRACT

The ability of animals to adapt to their changing environment will depend in part on shifts in their ranging patterns, but when and why individuals choose to move requires detailed understanding of their decision-making processes. We develop a simple decision-making model accounting for resource availability in habitually used ranges. We suggest that disparities between model predictions and animal tracking data indicate additional factors influencing movement decisions, which may be identified given detailed system-specific knowledge. The model was evaluated using movement data from satellite-tracked elephants (*Loxodonta africana*) inhabiting the Amboseli Ecosystem in Kenya, moving from savannah areas with low quality but constant resource availability, to areas with temporally-constrained higher nutrient availability. Overall, the model fit the data well: there was a good correlation between predicted and observed locations for the combined data from all elephants, but variation between individuals in how well the model fits. For those elephants where model predictions were less successful, additional factors likely to affect movement decisions, reproduction, anthropogenic threats, memory and perception, are suggested. This protocol for building and testing decision-making models should contribute to success in attempts to preserve sufficient space for large herbivores in their increasingly human-dominated ecosystems.

## KEYWORDS

Decision-making; Large herbivores; Movement ecology; NDVI; Optimal foraging; Remote sensing

## INTRODUCTION

In the face of a changing climate and the rapid conversion of natural habitats to human-dominated landscapes, the future of many species will depend on their ability to adapt to new circumstances. Adaptation may involve behavioural changes or innovations (Sol, Duncan, Blackburn, Cassey, & Lefebvre, 2005), but geographical shifts in a species range (Laidre et al., 2018) or in how a species moves through its existing range (Olden, Schooley, Monroe, & Poff, 2004; Tucker et al., 2018) may

allow it to cope with novel environmental conditions or constraints. Range shifts require animals to make the decision to relocate from one site to another and have often been predicted through the use of simple decision models (Bastille-Rousseau, Douglas-Hamilton, Blake, Northrup, & Wittemyer, 2018). Whilst these shifts are largely guided by suitable climatic conditions and the provision of sufficient food and water, movement decisions are also likely mediated by other factors. The predicted shifts can be incorporated into conservation and management strategies for species of concern, but are unlikely to be accurate unless the full range of factors influencing movement can be taken into account.

Here we present a means of identifying factors other than simple resource-requirements which mediate movement decisions for large herbivores negotiating heterogeneous landscapes. Our method uses disparities between a simple model of resource-driven decision making and animal tracking data to indicate the need to incorporate other factors that affect movement decisions. Given a detailed understanding of the study system, these factors can be identified using local knowledge (see also Bastille-Rousseau et al., 2018). Models were developed and evaluated using elephants (*Loxodonta africana*) inhabiting the Amboseli Basin in Kenya as a case study. Ensuring a future for elephants in this increasingly human-dominated landscape hinges on understanding how and why elephants use the landscape through space and over time.

Long-term monitoring of elephant populations and remote tracking studies have greatly improved our understanding of general patterns in elephant ranging behaviour (e.g. Wall et al. 2013). Elephants show large-scale, purposeful space use (Polansky, Kilian, & Wittemyer, 2015) and so require large areas over which to roam to access heterogeneously distributed resources (e.g. Leuthold 1977; Lindeque & Lindeque 1991; Thouless 1996; Blake et al. 2003; Birkett et al. 2012). It is also clear that elephants do not use the space available to them at random. Instead, elephants generally shift ranges seasonally (Leggett, 2006; Loarie, Van Aarde, & Pimm, 2009; Western & Lindsay, 1984), searching for water (Chamaille-Jammes & Valeix, 2007; de Beer & van Aarde, 2008; Redfern, Grant, Biggs, & Getz,

2015) and the highest quality vegetation (Bohrer, Beck, Ngene, Skidmore, & Douglas-Hamilton, 2014; Loarie et al., 2009). This results in dynamic habitat and food type preferences (Cerling et al., 2006; Loarie et al., 2009; G. Shannon, Page, Slotow, & Duffy, 2006).

We developed a simple decision-making model which tracked the food available within habitually-used ranges to satellite-tracked individuals from five family groups (representing the movements of over 220 elephants in the Amboseli population), whilst taking into account the daily need of individuals for water. The individuals had complete perceptual knowledge of resource availability in their home range and the model assumed that, when water availability allowed, individuals moved to maximise their nutrient intake rate and subsequent fitness (Okello, Njumbi, Kiringe, & Isiiche, 2015), as in optimal foraging theory (Bastille-Rousseau et al., 2017; Roever, van Aarde, & Chase, 2013; Stephens & Krebs, 1986; Vasconcelos, Fortes, & Kacelnik, 2017). We assumed that individuals make daily choices whether to relocate to an alternative location or to stay in the current location (Petit & Bon, 2010). Where model predictions did not match those of tracking data, we used detailed knowledge of the elephants and ecosystem to identify additional factors, such as physiological or social needs, which depend inter alia on an individual's sex, age, reproductive status and body condition (Lindsay, 2011).

## **MATERIALS AND METHODS**

### **Case study site information**

The study was carried out in the Amboseli basin (bounding coordinates: -2.02N, -3.28S, 38.03E, 36.67W), an area of approximately 8000km<sup>2</sup> straddling the border of southern Kenya and northern Tanzania, comprising the central Amboseli National Park (392km<sup>2</sup>; ANP hereafter) and surrounding landscape (Croze & Lindsay, 2011; see Supplementary Information). The habitat consists of semi-arid savannah that responds seasonally to highly variable rainfall. Rain predominantly occurs in two seasons; a short wet season (November and December) and the long rains (March through May; Altmann, Alberts, Altmann, & Roy, 2002). During dry seasons, a series of swamps in the central basin

provide the primary source of water, fed by groundwater flow from the slopes of Kilimanjaro in the south (Croze & Lindsay, 2011), though other perennial sources occur in the Selenkay Conservancy and in Kitenden, in the form of a borehole and two natural springs. Wildlife concentrates in the central basin during dry seasons and disperses following rain (Western, 1975). The Amboseli basin is home to around 1670 individually known and monitored elephants (Lee, Bussiere, Webber, Poole, & Moss, 2013). Unlike most other African populations, the Amboseli elephants have been relatively undisturbed by human activities. However, recent human population growth in Kenya and a lifestyle shift from nomadic pastoralism to sedentary farming (Western, Groom, & Worden, 2009) presents growing challenges for the persistence of Amboseli's wildlife.

#### **Animal tracking data**

GPS-GSM collars (@ Savannah Tracking) were fitted in July 2011 to five adult female elephants (Ida, Lobelia, Maureen, Vicky and Willow) from different family groups, representing locations of over 220 individuals (because families forage as units). Habitat heterogeneity around the central protected area affects ranging strategies and reproductive success, so not all dispersal directions are equal for Amboseli elephants. Target females were selected based on more than 40 years of observations to represent the known diversity in dispersal areas and foraging opportunities in Amboseli (which are largely inherited through generations unless disrupted by human disturbance; Croze & Moss, 2011; Moss, 1988) and to minimise risks to target elephants, family members and staff. Individual females were selected according to ethical and safety criteria, minimising the disturbance of each intervention. Target females were 1) not matriarchs, 2) without a calf aged <2 years and 3) closely related to matriarchs so shared movement patterns. Matriarchs were not collared due to the drastic potential impact of matriarch loss on families given the small but present risks of immobilisation, and the greater ease of managing non-target family members during immobilisations when they had a safe rallying point around their matriarch. Collars were fitted under the authority of the Kenya Wildlife Service, the Kenyan body regulating interactions with elephants, and with research clearance

to Amboseli Trust for Elephants from the National Commission for Science, Technology and Innovation (NACOSTI/P/15/9605/5732).

GPS fixes were recorded at hourly intervals for roughly 12 months, giving in sum 43,728 location fixes. Collar data was summarised into daily presence or absence from ANP. Given the reliance of elephant families on water, an elephant was considered present in ANP on any day in which distance from the swamp was zero at any time during that day. Conversely if the distance from the swamp was never zero, the elephant was considered that day to have dispersed from ANP. The dispersal area of each female was identified as the 95% kernel density estimates (ArcMap; ESRI 2017) of her locations outside the park boundary.

#### **Resource-driven movement model**

We developed a profitability index to indicate the resource availability of each dispersal area and of ANP, while taking into account the daily need of female elephants for water (Fig. 1). Water was considered essential and elephants were only able to move to areas where water was available. The swamp edge habitat alone was used to indicate profitability for ANP, as elephants consistently return to the park to feed on the reliable and abundant swamp edge vegetation, as well as drink. Swamp edge was therefore used as representative of the resources drawing elephants back to ANP from their dispersal areas.

#### *Vegetation quantity*

Data on vegetation quantity were acquired using the Normalised Difference Vegetation Index (NDVI), which exploits the marked difference in reflectance in red and near infra-red wavelengths characteristic of healthy green vegetation (Huete et al., 2002). We used 16-day composite values of NDVI retrieved from Terra-MODIS (Moderate Resolution Imaging Spectroradiometer) to infer time-specific values of vegetation quantity. Specifically we used the MOD13Q1 product accessed via the Oak Ridge National Laboratory Web Service (Vannan, Cook, Pan, & Wilson, 2011). Median NDVI values

were calculated for each individual's dispersal area (i.e. 95% kernel density estimates outside ANP) and swamp edge for each 16-day interval. Data were filtered using the MOD13Q1 QA flags such that only 'good' quality NDVI observations were used in our calculations. For extended remote sensing methodology, see Supplementary Information.

#### *Vegetation quality*

Crude protein is an important limiting factor for herbivores inhabiting savannah ecosystems (Sinclair, 1975) and we therefore assessed vegetation quality by its protein content (%). The diet of Amboseli elephants is dominated by grasses, so we used grass protein content as our measure of vegetation quality. Protein content varies seasonally, peaking during the green-up of vegetation following the onset of rains (Georgiadis & McNaughton, 1990; Lindsay, 1994). Here we estimate protein content depending on whether or not it rained in the previous month. Lindsay (1994) measured the protein content of Amboseli swamp edge vegetation and rainfall throughout the course of a year. From these data, we assigned protein content of swamp edge vegetation as 11.8% or 8.4% depending on whether or not it rained in the preceding month. Georgiadis & McNaughton (1990) collected similar measurements in the broader Amboseli basin outside the Amboseli swamps and found protein contents of 23.0% during the green-up following rains and 10.0% in the subsequent drying phases. We used the figure of 23.0% if the change in NDVI was positive, indicating green-up in the month after rains, at all other times 10.0%.

#### *Water availability*

Permanent water sources were available in ANP and in the dispersal areas of Vicky (Selenkay), and Ida and Lobelia (Kitenden). We deemed water available year-round in these areas. We used daily measures of rainfall from the rainfall gauge in the Amboseli Elephant Research Camp (-2.679S, 37.267E) to indicate rainfall across the entire Amboseli basin since NDVI fluctuations across the ecosystem are generally synchronous. Rainfall contributed to surface water availability in all areas and so water was considered available across the entire basin for seven days following rains.

## 179 *Movement-decision model fit*

180 We assumed that if individual movement behaviour was governed by resource availability, individuals  
181 should move to maximise profitability throughout the year. Therefore, when ANP profitability was  
182 greater than that of the dispersal area, the individual should be present in ANP on that day, and vice  
183 versa. If profitability for the two locations was very close (difference < 0.3) no prediction was made as  
184 to which provided the optimal foraging location.

185 The daily absence or presence of the elephants as predicted by profitability was compared to actual  
186 absence or presence indicated by the collar data. The correlation between predicted and observed  
187 presences and absences was calculated as a  $\phi$  statistic (Conover, 1971).  $\phi$  is the equivalent of  
188 Pearson's correlation that is applicable to binary data.  $\phi$  values were tested for significance using chi-  
189 square with 1 degree of freedom.

## 190 **RESULTS**

### 191 **Ranging behaviour**

192 The ranging behaviour of the five collared elephants over a 12-month period is illustrated in Fig. 2. Ida  
193 and Lobelia spent most of their time (c. 85%) in ANP, primarily in the southeast around the southern  
194 tip of the eastern swamp (Longinye; see Supplementary Information for detailed park map). From  
195 there both elephants regularly moved southwest into the Kitenden region and occasionally east to the  
196 Kimana Sanctuary. Maureen, Vicky and Willow spent about half their time in the park. Maureen used  
197 the eastern swamp and the area around and including the western swamp (Longolong). From there,  
198 she dispersed south through the Kitirua Conservancy and then southwest across the foothills of  
199 Kilimanjaro into Tanzania. Vicky and Willow also used the eastern swamp, but additionally used the  
200 northern tip of the central swamp (Enkongo Narok). Within the park Vicky frequented the north and  
201 dispersed north to spend much time in the Selenkay Conservancy. Willow by contrast used the west  
202 of the park from which she dispersed northwest to the Meshanani region.



## 203    **Profitability**

204    Profitabilities calculated for the swamp edge and dispersal areas (Kitenden, Kitirua-Tanzania, Selenkay  
205    and Meshanani) captured spatial and temporal variation (Fig. 3). Temporally, profitability generally  
206    increased following the onset of the rains and declined as the rains subsided. Spatially, areas differed  
207    in the precise timing and extent of increases and decreases in profitability. ANP profitability varied  
208    less than the profitability of the dispersal areas because in the dry seasons the swamps retained  
209    abundant green vegetation but did not experience the dramatic increase in protein content seen in  
210    dispersal area vegetation following rain. As a result, there was temporal variation in whether  
211    profitability was higher in ANP or in the dispersal area, predicting switches in the optimal foraging  
212    location between ANP and the dispersal areas. Generally, profitability indices predicted that elephants  
213    should be present in the national park during August, September and October 2011, January and  
214    February 2012 and from May 2012 onwards. At other times they were predicted to move out to their  
215    family dispersal areas.

## 216    **Movement-decision model fit**

217    Overall the model fitted the data well: there was good correlation between predicted and observed  
218    presences and absences for the combined data from all five elephants ( $\phi = 0.37$ ,  $p < 0.001$ ), but there  
219    was variation between elephants in how well the model fitted (Breslow-Day test,  $\chi^2_4 = 67.4$ ,  $p < 0.001$ ,  
220    Fig. 4). The locations of Vicky were well predicted by the model ( $\phi = 0.60$ ,  $p < 0.001$ , Fig. 4) though the  
221    model was unable to predict Vicky's brief excursions between August and October 2011, nor her  
222    absence from the park in May to June 2012. The model was moderately successful in predicting the  
223    locations of Ida, Lobelia, Maureen and Willow ( $\phi = 0.42, 0.27, 0.39$  and  $0.46$  respectively,  $p < 0.001$  for  
224    each), but failed to predict the regular presence of Ida and Lobelia in ANP throughout both wet  
225    seasons. Maureen was regularly absent between August and October 2011 and in June and July 2012  
226    when the model assumed presence throughout. The timings of Maureen's major excursions from the  
227    park were also slightly ahead of those predicted by the model. Willow's movements were well

predicted with the exception of brief excursions during August and September 2011 and of her continued absence from the park during June and July 2012.

## DISCUSSION

By modelling the profitability of the traditional foraging areas of each of the five elephant families while taking account of their daily need for water, we predicted the timings of their shifts in optimal locations. The shift timings were similar despite the different dispersal areas used by the collared elephants (Fig. 3). In general, ANP offered the highest rate of nutrient intake during the dry seasons until the onset of rains, at which point the dispersal areas provided better foraging locations provided that water was available. While the model fitted the data well overall, there was variation among elephants in the ability of the model to correctly predict park absence and presence. Predictions and observations were significantly correlated for all five individuals, but while the correlation was good for Vicky, it is likely that other factors also influenced the movement decisions of Ida, Lobelia, Maureen and Willow. Given an in-depth knowledge of the study system (AERP long term data), we now attempt to identify these factors and suggest how they can be incorporated into future movement models for elephants.

The sustained residency of Ida and Lobelia in ANP is likely due to the fact that both females gave birth during the 2012 short wet season (January-February). Elephants usually seek safe areas as parturition approaches and for some time after as new-born calves are vulnerable to disturbance and predation (Ruggiero, 1991). For family units experiencing recent parturition events, the motivation to remain in ANP increases because it is perceived as a relatively safe area. During the period of sampling, Ida and Lobelia's dispersal areas of Kitenden and Kimana were relatively high risk because a number of elephants were killed or injured as a result of human interactions (Big Life Foundation/AERP long term data). In elephants, risks in the form of anthropogenic threats are known to deter long-term elephant habitat use (Roever et al., 2013) and alter the daily locations of resting sites (Wittemyer, Keating, Vollrath, & Douglas-Hamilton, 2017). Risks affect animal decision making more widely when animals

sacrifice optimum nutrient intake to minimise the risks of predation (Barnier et al., 2014; Bastille-Rousseau et al., 2017). Including the risks associated with human-elephant interactions should improve the model fit. Relative risk could be measured in terms of the number of negative human-elephant interactions (injuries or mortalities) over a specified period, land use type (pastoralist vs. agricultural vs. wildlife zones) or human density in an area (which is also a function of land use type). Over the longer-term, periods of drought could also be used as an indirect predictor of human-associated risk, because competition for water sources and high-quality food patches increases contact between humans and elephants (Chiyo, Cochrane, Naughton, & Basuta, 2005) and the economic threats of livestock loss erode tolerance for wildlife (Western, Nightingale, Sipitiek, Mose, & Kamiti, pers. comm.).

Behavioural differences between conspecifics can be viewed as evidence of personality; consistent differences in the behavioural responses of individuals across various spatial and temporal contexts (Beekman & Jordan, 2017). Here therefore, we may alternatively consider the residency of Ida and Lobelia could be due to their 'wary' personalities, whilst Vicky is more 'bold' (Jolles, Boogert, Sridhar, Couzin, & Manica, 2017). Personality has been reliably demonstrated for elephants (Lee & Moss, 2012; Selmann, Helle, Adams, Mar, & Lahdenperä, 2018) but we have yet to systematically sample the subject families and can only speculate at this point. However, our long-term observations suggest that personality is shaped by family members (particularly the matriarch) and by experiences, especially early in life. We can propose that, as seen in many other bird, fish and mammal species (Weiss, 2018), personality may both shape and be shaped by exposure to risk.

Maureen displayed much lower use of ANP than predicted by the model, suggesting she was less reliant on the swamps for dry season water as she had access to an alternative water source in the Kitirua Conservancy. Elephants share this water point with livestock and people, often at high concentrations, but this area also showed high tolerance for wildlife during the study period. The model was also unable to predict the timing of Maureen's location shift, consistently predicting a later

278 dispersal than seen in the collar data. The dispersal area used by Maureen in Tanzania is characterised  
279 by mature Acacia woodlands rather than the open bushed savannah associated with the majority of  
280 the ecosystem. Acacia flower ahead of the onset of rains and Maureen's early excursions may  
281 coincide with this phenological event, rather than the NDVI observed green-up. Flowering in  
282 bushlands might be associated with higher sugar and other nutrient transport in Acacia stems, which  
283 are desirable elephant foods (Lindsay, 1994). Incorporating nutrient fluxes that are independent of  
284 vegetation green-up will require both phenological monitoring of the timing of flowering and  
285 quantifying the associated nutrient benefits. The remotely sensed measures of greenness used here  
286 are insensitive to the spectral changes associated with flowering, thus suitable alternative remote  
287 sensing techniques are needed for a more comprehensive accounting for foraging movements.

288 While we can suggest other factors that operate in combination with nutrient intake to affect  
289 movement decisions, the behavioural mechanisms underlying movement patterns remain poorly  
290 understood (Bolger, Newmark, Morrison, & Doak, 2008); indeed whether movement decisions are  
291 guided by animals' perceptions of current environmental conditions or predictions based on memory  
292 is uncertain. The capacity of elephants to remember the spatial locations of out-of-sight individuals  
293 (Bates et al., 2008) and the purposeful movement to water resources beyond the senses of sight or  
294 smell (Polansky et al., 2015) shows that elephants have a keen spatial memory. Despite this, our data  
295 show that decision points (Polansky et al., 2015) coincide with the onset of the rains, suggesting that  
296 the onset of rains may be used as a key environmental cue to instigate range shifts (Holdo, Holt, &  
297 Fryxell, 2009; Prins, 1996). Also relevant to this question are several brief excursions by Vicky and  
298 Willow to their respective dispersal areas prior to their longer-term range shifts. This 'scouting'  
299 behaviour (Bracis & Mueller, 2017) suggests sampling of the conditions at their destination before  
300 committing to the final range shift. Both of these factors suggest perception-guided movement.  
301 However, rainfall across the Amboseli basin is perhaps more temporally and spatially variable than  
302 accounted for here, meaning elephants cannot rely on the same rules each year. Memory alone is  
303 likely to be less effective in such a patchy environment, and so individuals will use both immediate

perception and memory when making movement decisions. Although memory is important in defining the destination of movements based on previous experiences of improved nutrient intake rate, perception is crucial in remaining flexible to local environmental changes in climatic conditions and risk.

We have introduced a simple decision-making model of the resource-driven factors affecting key decisions of where individuals go and when. Where individuals fit the model well, a key benefit of movement over an ecosystem is indicated, which is to maximise nutrient intake rates. That individuals do not perfectly fit the model suggests that other factors need to be included in combination with nutrient intake rates. High residency of elephants that disperse to the eastern part of the ecosystem suggests that individuals and families make movement decisions based on trade-offs between the resource-related benefits of dispersal and the associated risks (see also Chiyo et al., 2014). For *Ida* and *Lobelia*, it is hard to determine whether perceived risks resulted from human-elephant interactions, the dangers involved with moving new-born calves over long distances, or the limited travel speed of those vulnerable calves. More data characterising risks and associated movements are important (Nielsen, Stenhouse, & Boyce, 2006; Roever et al., 2013) and necessary before these factors can be reliably included in an enhanced model of movement decision making. Male elephants were not considered here due to differences in resource requirements. Males are less reliant on water than females and forage less selectively, prioritising quantity over quality (Graeme Shannon, Page, Duffy, & Slotow, 2006). When in musth, male elephant movement is largely focused on the pursuit of oestrous females, with little consideration for food (Poole, 1987). Previous work in Amboseli has demonstrated the importance of both food (NDVI) and social needs (Chiyo et al. 2014) on male ranging, so we expect that our model could be extended to non-musth males but would need further modification to take into account the energy demands of musth, which is a topic for future study. Further development of our model will ultimately provide a basis for robust prediction of elephant movements under a variety of environmental and physiological conditions.

329 We argue that simple resource-driven movement models based on easily accessible resource  
330 availability data in combination with animal tracking studies will be useful in identifying additional  
331 features influencing movements in well-studied systems, where anecdotal evidence can add to  
332 understanding movement decisions. This will allow conservationists and wildlife managers to better  
333 understand how large herbivore movements will respond to management scenarios (e.g. fences, new  
334 roads) and future environmental changes (e.g. shifting rainfall patterns), and in this way can help  
335 manage conflicts of interest between humans and wildlife and ensure sufficient space for wildlife.

336

## 337 ACKNOWLEDGEMENTS

338 We thank the Government of Kenya for permission to conduct research, our field team (T. Manor, N.  
339 Njiraini, K. Sayialel, M. Sowers) for monitoring, the Kenya Wildlife Service, the Maasai community who  
340 share land with elephants, and our many generous donors over the duration of the project. M.  
341 Sowers was supported by a Fullbright grant. This work was supported by the Natural Environment  
342 Research Council (grant number NE/L002566/1).

343

## 344 REFERENCES

- 345 Altmann, J., Alberts, S. C., Altmann, S. A., & Roy, S. B. (2002). Dramatic change in local climate patterns  
346 in the Amboseli basin, Kenya. *African Journal of Ecology*, 40(3), 248–251.  
347 <https://doi.org/10.1046/j.1365-2028.2002.00366.x>
- 348 Barnier, F., Valeix, M., Duncan, P., Chammille-Jammes, S., Barre, P., Loveridge, A. J., ... Fritz, H. (2014).  
349 Diet quality in a wild grazer declines under the threat of an ambush predator. *Proc Biol Sci*, 281.  
350 <https://doi.org/10.1098/rspb.2014.0446>
- 351 Bastille-Rousseau, G., Douglas-Hamilton, I., Blake, S., Northrup, J. M., & Wittemyer, G. (2018).  
352 *Applying network theory to animal movements to identify properties of landscape space use.*

353 *Ecological Applications*. <https://doi.org/10.1002/eap.1697>

354 Bastille-Rousseau, G., Murray, D. L., Schaefer, J. A., Lewis, M. A., Mahoney, S. P., & Potts, J. R. (2017).  
355 Spatial scales of habitat selection decisions: Implications for telemetry-based movement  
356 modelling. *Ecography*. <https://doi.org/10.1111/ecog.02655>

357 Bates, L. A., Sayialel, K. N., Njiraini, N. W., Poole, J. H., Moss, C. J., & Byrne, R. W. (2008). African  
358 elephants have expectations about the locations of out-of-sight family members. *Biol. Lett*, 4,  
359 34–36. <https://doi.org/10.1098/rsbl.2007.0529>

360 Beekman, M., & Jordan, L. A. (2017). Does the field of animal personality provide any new insights for  
361 behavioral ecology? *Behavioral Ecology*, 28(3), 617–623.  
362 <https://doi.org/10.1093/beheco/axx022>

363 Birkett, P. J., Vanak, A. T., Muggeo, V. M. R., Ferreira, S. M., & Slotow, R. (2012). Animal perception of  
364 seasonal thresholds: Changes in elephant movement in relation to rainfall patterns. *PLoS ONE*,  
365 7(6), 1–8. <https://doi.org/10.1371/journal.pone.0038363>

366 Blake, S., Bouché, P., Rasmussen, H., Orlando, A., & Douglas-Hamilton, I. (2003). *The Last Sahelian*  
367 *Elephants Ranging Behavior, Population Status and Recent History of the Desert Elephants of*  
368 *Mali*. <https://doi.org/11858/00-001M-0000-002D-0B5D-1>

369 Bohrer, G., Beck, P. S., Ngene, S. M., Skidmore, A. K., & Douglas-Hamilton, I. (2014). Elephant  
370 movement closely tracks precipitation-driven vegetation dynamics in a Kenyan forest-savanna  
371 landscape. *Movement Ecology*, 2(2), 1–12. <https://doi.org/10.1186/2051-3933-2-2>

372 Bolger, D. T., Newmark, W. D., Morrison, T. A., & Doak, D. F. (2008). The need for integrative  
373 approaches to understand and conserve migratory ungulates. *Ecology Letters*, 11, 63–77.  
374 <https://doi.org/10.1111/j.1461-0248.2007.01109.x>

375 Bracis, C., & Mueller, T. (2017). Memory, not just perception, plays an important role in terrestrial  
376 mammalian migration. *Proceedings of the Royal Society B*, 284.

377 <https://doi.org/10.1098/rspb.2017.0449>

378 Cerling, T. E., Wittemyer, G., Rasmussen, H. B., Vollrath, F., Cerling, C. E., Robinson, T. J., & Douglas-  
379 Hamilton, I. (2006). Stable isotopes in elephant hair document migration patterns and diet  
380 changes. *PNAS*, 103(2), 371–373. <https://doi.org/10.1073/pnas.0509606102>

381 Chamaille-Jammes, S., & Valeix, M. (2007). Managing heterogeneity in elephant distribution:  
382 interactions between elephant population density and surface-water availability. *Journal of*  
383 *Applied Ecology*, 44, 625–633. <https://doi.org/10.1111/j.1365-2664.2007.01300.x>

384 Chiyo, P. I., Cochrane, E. P., Naughton, L., & Basuta, G. I. (2005). Temporal patterns of crop raiding by  
385 elephants: A response to changes in forage quality or crop availability? *African Journal of*  
386 *Ecology*, 43, 48–55. <https://doi.org/10.1111/j.1365-2028.2004.00544.x>

387 Chiyo, P. I., Wilson, J. W., Archie, E. A., Lee, P. C., Moss, C. J., & Alberts, S. C. (2014). The influence of  
388 forage, protected areas, and mating prospects on grouping patterns of male elephants.  
389 *Behavioural Ecology*, 0(0), 1–11. <https://doi.org/10.1093/beheco/aru152>

390 Conover, W. J. (1971). *Practical Nonparametric Statistics*. New York: John Wiley & Sons Inc.

391 Croze, H., & Lindsay, W. K. (2011). Amboseli Ecosystem Context: Past and Present. In C. J. Moss, H.  
392 Croze, & P. C. Lee (Eds.), *The Amboseli Elephants: A Long-Term Perspective on a Long-Lived*  
393 *Mammal* (pp. 11–28). Chicago: University of Chicago Press.

394 Croze, H., & Moss, C. J. (2011). Patterns of Occupancy in Space and Time. In C. J. Moss, H. Croze, & P.  
395 C. Lee (Eds.), *The Amboseli Elephants: A Long-Term Perspective on a Long-Lived Mammal* (pp.  
396 89–105). Chicago: University of Chicago Press.

397 de Beer, Y., & van Aarde, R. J. (2008). Do landscape heterogeneity and water distribution explain  
398 aspects of elephant home range in southern Africa's arid savannas? *Journal of Arid*  
399 *Environments*, 72(11), 2017–2025. <https://doi.org/10.1016/j.jaridenv.2008.07.002>



400 Georgiadis, N. J., & McNaughton, S. J. (1990). Elemental and fibre contents of savanna grasses:  
 401 variation with grazing, soil type, season and species. *Journal of Applied Ecology*, 27(2), 623–634.  
 402 <https://doi.org/10.2307/2404307>

403 Holdo, R. M., Holt, R. D., & Fryxell, J. M. (2009). Opposing Rainfall and Plant Nutritional Gradients Best  
 404 Explain the Wildebeest Migration in the Serengeti. *The American Naturalist April*, 173(4).  
 405 <https://doi.org/10.1086/597229>

406 Huete, A., Didan, K., Miura, T., Rodriguez, E. P., Gao, X., & Ferreira, L. G. (2002). Overview of the  
 407 radiometric and biophysical performance of the MODIS vegetation indices. *Remote Sensing of*  
 408 *Environment*, 83(1), 195–213. [https://doi.org/10.1016/S0034-4257\(02\)00096-2](https://doi.org/10.1016/S0034-4257(02)00096-2)

409 Jolles, J. W., Boogert, N. J., Sridhar, V. H., Couzin, I. D., & Manica, A. (2017). Consistent individual  
 410 differences drive collective behaviour and group functioning of schooling fish. *bioRxiv*, 131094,  
 411 1–18. <https://doi.org/10.1101/131094>

412 Laidre, K. L., Born, E. W., Atkinson, S. N., Wiig, Ø., Andersen, L. W., Lunn, N. J., ... Heagerty, P. (2018).  
 413 Range contraction and increasing isolation of a polar bear subpopulation in an era of sea-ice  
 414 loss. *Ecology and Evolution*, (November 2017), 2062–2075. <https://doi.org/10.1002/ece3.3809>

415 Lee, P. C., Bussiere, L. F., Webber, C. E., Poole, J. H., & Moss, C. J. (2013). Enduring consequences of  
 416 early experiences: 40 year effects on survival and success among African elephants (*Loxodonta*  
 417 *africana*). *Biology Letters*, 9(2). <https://doi.org/10.1098/rsbl.2013.0011>

418 Lee, P. C., & Moss, C. J. (2012). Wild female African elephants (*Loxodonta africana*) exhibit personality  
 419 traits of leadership and social integration. *Journal of Comparative Psychology*, 126(3), 224–232.  
 420 <https://doi.org/10.1037/a0026566>

421 Leggett, K. E. A. (2006). Home range and seasonal movement of elephants in the Kunene Region,  
 422 northwestern Namibia. *African Zoology*, 41(1), 17–36.  
 423 <https://doi.org/10.1080/15627020.2006.11407332>

424 Leuthold, W. (1977). Spatial organization and strategy of habitat utilization of elephants in Tsavo  
 425 National Park, Kenya. *Zeitschrift Fur Saugetierkunde*, 42, 358–379.

426 Lindeque, M., & Lindeque, P. M. (1991). Satellite tracking of elephants in northwestern Namibia.  
 427 *Journal of African Ecology*, 29(3), 196–206. <https://doi.org/10.1111/j.1365-2028.1991.tb01002.x>

428 Lindsay, W. K. (1994). *Feeding ecology and population demography of African elephants in Amboseli,*  
 429 *Kenya (PhD)*. University of Cambridge.

430 Lindsay, W. K. (2011). Habitat Use, Diet Choice, and Nutritional Status in Female and Male Amboseli  
 431 Elephants. In C. J. Moss, H. Croze, & P. C. Lee (Eds.), *The Amboseli Elephants: A Long-Term*  
 432 *Perspective on a Long-Lived Mammal* (pp. 51–73). The University of Chicago Press.

433 Loarie, S. R., Van Aarde, R. J., & Pimm, S. L. (2009). Elephant seasonal vegetation preferences across  
 434 dry and wet savannas. *Biological Conservation*, 142(12), 3099–3107.  
 435 <https://doi.org/10.1016/j.biocon.2009.08.021>

436 Moss, C. J. (1988). *Elephant Memories: Thirteen Years in the Life of an Elephant Family*. Chicago:  
 437 University of Chicago Press.

438 Nielsen, S. E., Stenhouse, G. B., & Boyce, M. S. (2006). A habitat-based framework for grizzly bear  
 439 conservation in Alberta. *Biological Conservation*, 130(2), 217–229.  
 440 <https://doi.org/10.1016/j.biocon.2005.12.016>

441 Okello, M. M., Njumbi, S. J., Kiringe, J. W., & Isiiche, J. (2015). Habitat use and preference by the  
 442 African elephant outside of the protected area, and management implications in the Amboseli  
 443 Landscape, Kenya. *International Journal of Biodiversity and Conservation*, 7(3), 211–236.  
 444 <https://doi.org/10.5897/IJBC2014.0795>

445 Olden, J. D., Schooley, R. L., Monroe, J. B., & Poff, N. L. (2004). Context-dependent perceptual ranges  
 446 and their relevance to animal movements in landscapes. *Journal of Animal Ecology*.  
 447 <https://doi.org/10.1111/j.0021-8790.2004.00889.x>

448 Petit, O., & Bon, R. (2010). Decision-making processes: The case of collective movements. *Behavioural*  
 449 *Processes*, 84, 635–647. <https://doi.org/10.1016/j.beproc.2010.04.009>

450 Polansky, L., Kilian, W., & Wittemyer, G. (2015). Elucidating the significance of spatial memory on  
 451 movement decisions by African savannah elephants using state-space models. *Proceedings of*  
 452 *the Royal Society B: Biological Sciences*, 282. <https://doi.org/10.1098/rspb.2014.3042>

453 Poole, J. H. (1987). Rutting behaviour in African elephants: the phenomenon of musth. *Behaviour*,  
 454 102(34), 283–316. <https://doi.org/10.1163/156853986X00171>

455 Prins, H. H. T. (1996). *Ecology and Behaviour of the African Buffalo*. Chapman & Hall.  
 456 <https://doi.org/10.1007/978-94-009-1527-5>

457 Redfern, J. V., Grant, R., Biggs, H., & Getz, W. M. (2015). Surface-Water Constraints on Herbivore  
 458 Foraging in the Kruger National Park , South Africa. *Ecology*, 84(8), 2092–2107.  
 459 <https://doi.org/10.1890/01-0625>

460 Roever, C. L., van Aarde, R. J., & Chase, M. J. (2013). Incorporating mortality into habitat selection to  
 461 identify secure and risky habitats for savannah elephants. *Biological Conservation*, 164, 98–106.  
 462 <https://doi.org/10.1016/j.biocon.2013.04.006>

463 Ruggiero, R. G. (1991). Opportunistic predation on elephant calves. *African Journal of Ecology*, 29(1),  
 464 86–89. <https://doi.org/10.1111/j.1365-2028.1991.tb00823.x>

465 Seltmann, M. W., Helle, S., Adams, M. J., Mar, K. U., & Lahdenperä, M. (2018). Evaluating the  
 466 personality structure of semi-captive Asian elephants living in their natural habitat. *Royal Society*  
 467 *Open Science*, 5. <https://doi.org/http://dx.doi.org/10.1098/rsos.172026>

468 Shannon, G., Page, B. R., Duffy, K. J., & Slotow, R. (2006). International Association for Ecology The  
 469 Role of Foraging Behaviour in the Sexual Segregation of the African Elephant. *Oecologia*, 150(2),  
 470 344–354. Retrieved from <http://www.jstor.org/stable/40210553>

471 Shannon, G., Page, B., Slotow, R., & Duffy, K. (2006). African elephant home range and habitat  
 472 selection in Pongola Game Reserve, South Africa. *African Zoology*, 41(1), 37–44.  
 473 <https://doi.org/10.1080/15627020.2006.11407333>

474 Sinclair, A. R. E. (1975). The Resource Limitation of Trophic Levels in Tropical Grassland Ecosystems.  
 475 *Source Journal of Animal Ecology*, 44(2), 497–520. <https://doi.org/10.2307/3608>

476 Sol, D., Duncan, R. P., Blackburn, T. M., Cassey, P., & Lefebvre, L. (2005). Big brains, enhanced  
 477 cognition, and response of birds to novel environments. *Proceedings of the National Academy of*  
 478 *Sciences*, 102(15), 5460–5465. <https://doi.org/10.1073/pnas.0408145102>

479 Stephens, D. W., & Krebs, J. R. (1986). *Foraging theory*. New Jersey: Princetown University Press.

480 Thouless, C. R. (1996). Home ranges and social organization of female elephants in northern Kenya.  
 481 *African Journal of Ecology*, 34(3), 284–297. <https://doi.org/10.1111/j.1365-2028.1996.tb00623.x>

482 Tucker, M. A., Böhning-Gaese, K., Fagan, W. F., Fryxell, J. M., Van Moorter, B., Alberts, S. C., ...  
 483 Mueller, T. (2018). Moving in the Anthropocene: Global reductions in terrestrial mammalian  
 484 movements. *Science*, 359, 466–469. Retrieved from  
 485 <http://science.sciencemag.org/content/sci/359/6374/466.full.pdf>

486 Vannan, S. K. S., Cook, R. B., Pan, J. Y., & Wilson, B. E. (2011). A SOAP Web Service for accessing  
 487 MODIS land product subsets. *Earth Science Informatics*, 4(2), 97–106.  
 488 <https://doi.org/10.1007/s12145-011-0079-2>

489 Vasconcelos, M., Fortes, I., & Kacelnik, A. (2017). On the structure and role of optimality models in the  
 490 study of behavior. In J. Call (Ed.), *APA Handbook of Comparative Psychology* (pp. 287–307).  
 491 Washington DC: American Psychological Association.

492 Wall, J., Wittemyer, G., Klinkenberg, B., Lemay, V., & Douglas-Hamilton, I. (2013). Characterizing  
 493 properties and drivers of long distance movements by elephants (*Loxodonta africana*) in the  
 494 Gourma, Mali. *Biological Conservation*, 157, 60–68.

<https://doi.org/10.1016/j.biocon.2012.07.019>

Weiss, A. (2018). Personality Traits: A View From the Animal Kingdom. *Journal of Personality*, 86(1), 12–22. <https://doi.org/10.1111/jopy.12310>

Western, D. (1975). Water availability and its influence on the structure and dynamics of a savannah large mammal community. *East African Wildlife Journal*, 13, 265–286. <https://doi.org/10.1111/j.1365-2028.1975.tb00139.x>

Western, D., Groom, R., & Worden, J. (2009). The impact of subdivision and sedentarization of pastoral lands on wildlife in an African savanna ecosystem. *Biological Conservation*, 142(11), 2538–2546. <https://doi.org/10.1016/j.biocon.2009.05.025>

Western, D., & Lindsay, W. K. (1984). Seasonal herd dynamics of a savanna elephant population. *African Journal of Ecology*, 22(4), 229–244. <https://doi.org/10.1111/j.1365-2028.1984.tb00699.x>

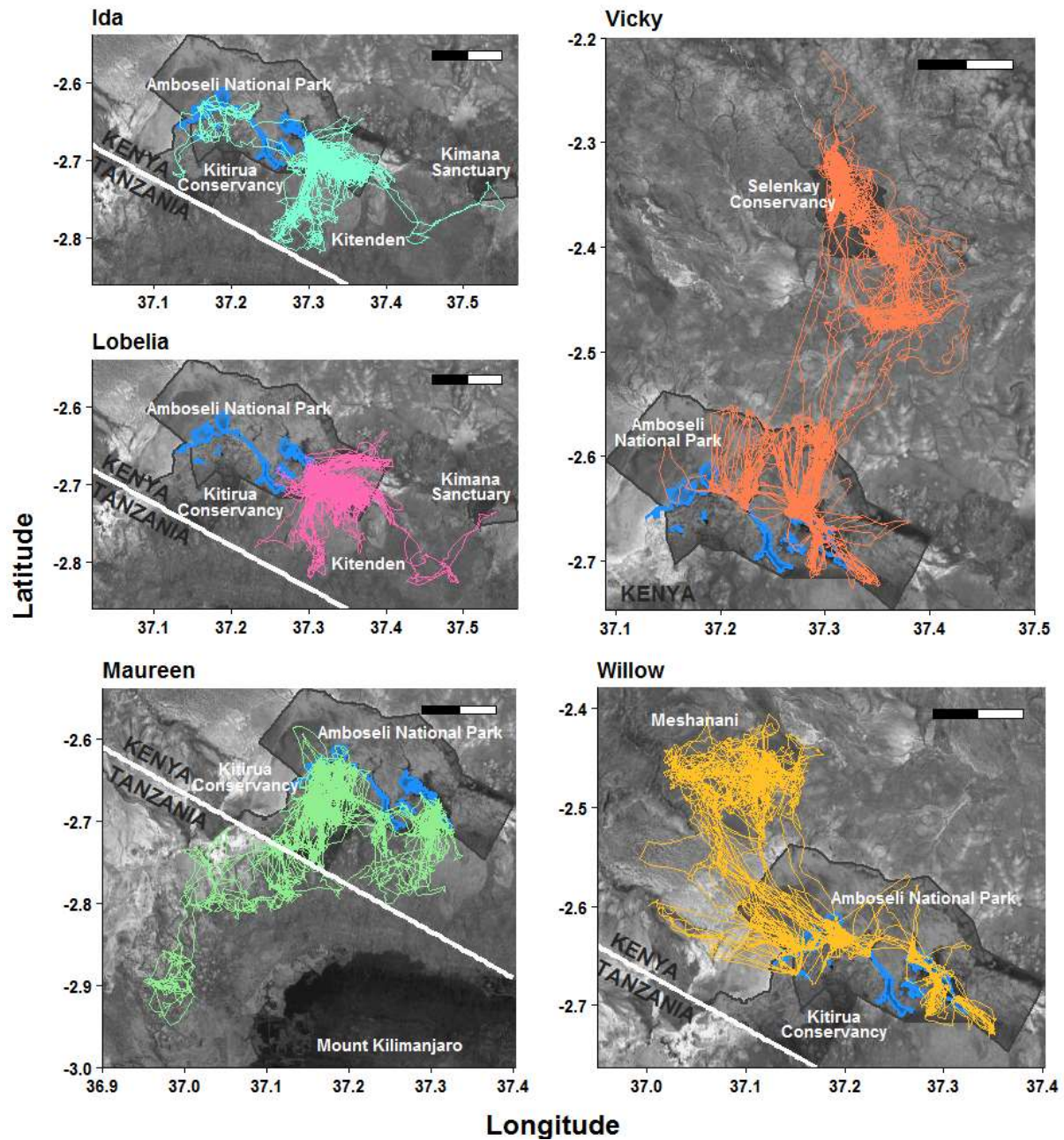
Wittemyer, G., Keating, L. M., Vollrath, F., & Douglas-Hamilton, I. (2017). Graph theory illustrates spatial and temporal features that structure elephant rest locations and reflect risk perception. *Ecography*, 40, 598–605. <https://doi.org/10.1111/ecog.02379>

## FIGURES

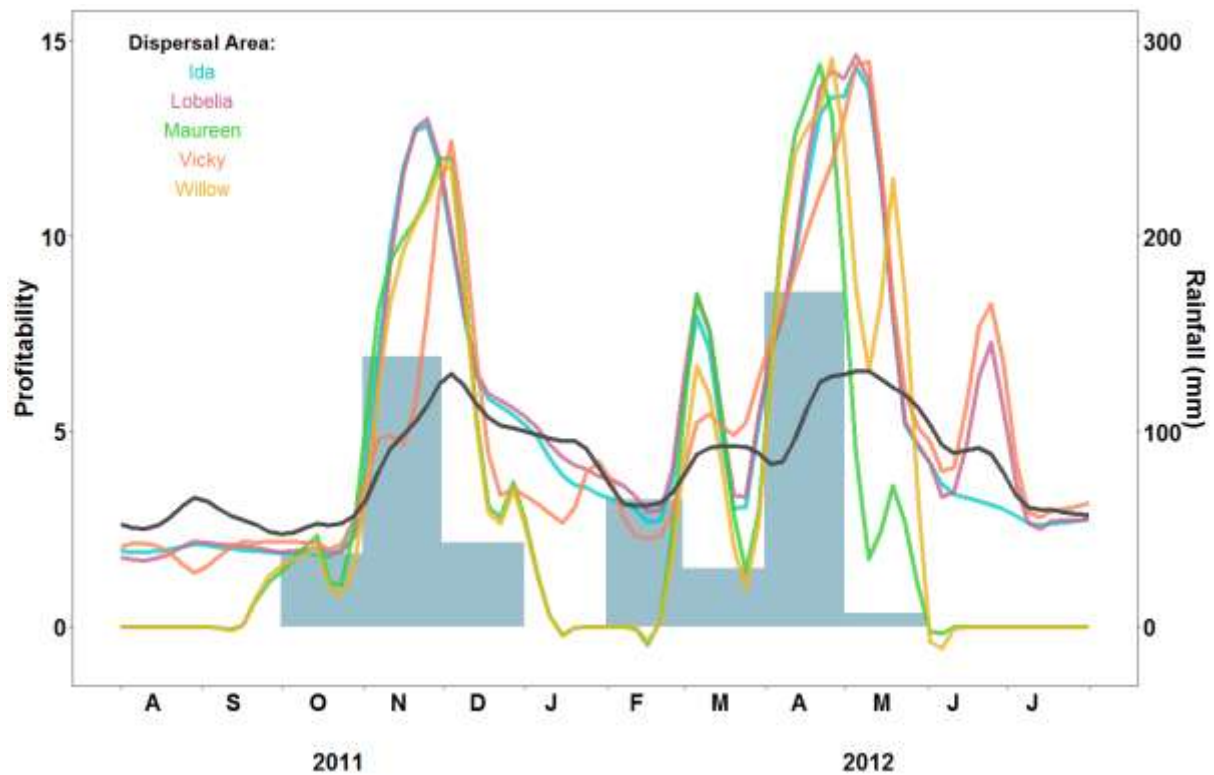
$$\text{Profitability of specified area} = \text{Quantity of vegetation} \times \text{Quality of vegetation} \times \text{Water availability}$$

Median NDVI      Protein content (%)      Binary

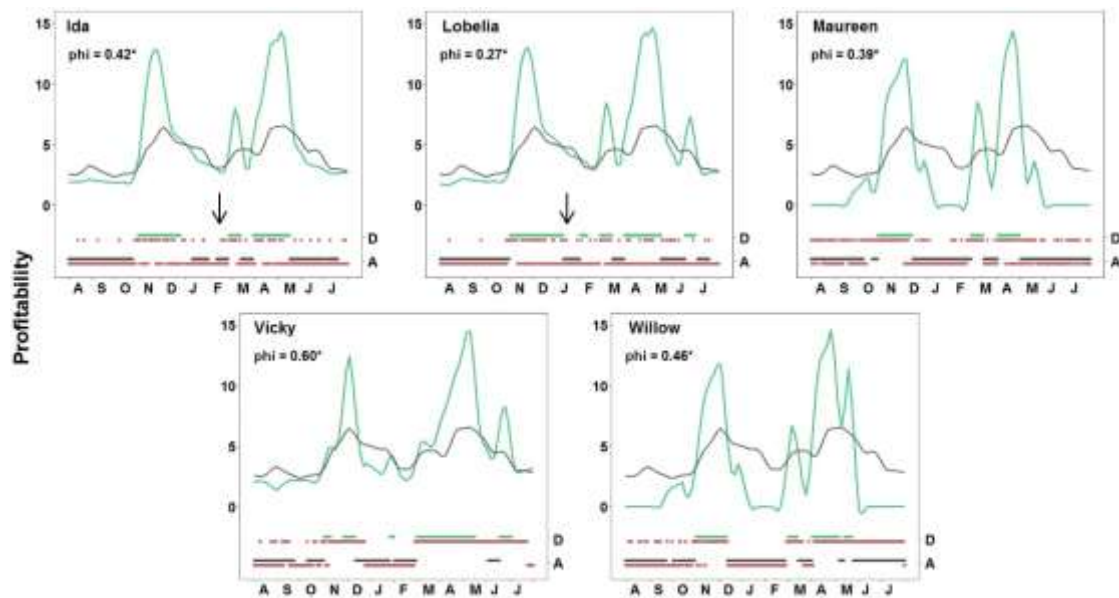
**Figure 1.** Profitability of each dispersal area and ANP, calculated daily. See Methods for definitions of vegetation quantity and quality. Water availability is a binary variable taking values of 1 or 0 depending on whether or not water is available in the specified area.



**Figure 2.** Tracks of five collared elephants displaying use of Amboseli National Park and dispersal areas (coloured lines: different colour for each female). Protected areas are indicated by grey boundaries, the international border between Kenya and Tanzania by the white line, and the central Amboseli swamps in blue. Scale bar represents 10km (divisions of 5km). Two females used the same dispersal area (Kitenden, southeast); other females ranged to the north, northwest and southwest when leaving ANP.



**Figure 3.** Response of profitability (lines: black = swamp edge, coloured = dispersal areas), the product of the quantity and quality of vegetation, and water availability, to monthly rainfall (bar: blue).



**Figure 4.** Profitabilities of swamp edge (black curve) and dispersal area (green curve) for each collared individual. Horizontal lines at bottom of each panel indicate model predictions (black, A = Amboseli National Park; green, D = dispersal area) and collar data (red). No predictions were made if

529     profitabilities were within 0.3 of each other. Arrows indicate parturition events for Ida and Lobelia. Phi  
530     coefficients indicate correlation between model predictions and collar data; \* indicates significance of  
531     this relationship (Chi-squared:  $p < 0.001$ ).